

Math 408

Will post the Exam I paper by
Friday 5:00 p.m. (Take home).

Due next. Friday noon.

No homework due next Friday.

$u^t u = I$ \rightarrow orthogonal matrix
 $\forall u$ is real, $u^t u = I = u u^t$

2.6 Remarks on real matrices

Theorem 2.6.1 If $A \in M_n(\mathbb{R})$, then A is orthogonally similar to block triangular matrix (A_{ij}) such that A_{jj} is either 1×1 or of the forms $\begin{pmatrix} a & b \\ -b & a \end{pmatrix}$. \checkmark

$$\begin{bmatrix} a, b & x & x \\ -b, a & & \\ 0 & \vdots & \vdots \end{bmatrix}$$

- The matrix is normal, i.e., $AA^t = A^t A$, if and only if all the off-diagonal blocks are zero.
- The matrix A orthogonal, i.e., $AA^t = I_n$, if and only if the 1×1 diagonal blocks have the form $[1]$ or $[-1]$, and the entries in the 2×2 diagonal blocks satisfy $a^2 + b^2 = 1$.
- The matrix A is symmetric, i.e., $A = A^t$, if and only if it is orthogonally similar to a real diagonal matrices.
- The matrix A is symmetric and satisfies $x^t A x \geq 0$ for all $x \in \mathbb{R}^n$ if and only if it is orthogonally similar to a nonnegative diagonal matrices.

Remark 2.6.2 Let $A \in M_n(\mathbb{R})$. Then $A = S + K$ where $S = (A + A^t)/2$ is symmetric and $K = (A - A^t)/2$ is skew-symmetric, i.e., $K^t = -K$.

- Note that $x^t K x = 0$ for all $x \in \mathbb{R}^n$. $\equiv (x^t K x)^t = x^t K^t x = x^t (-K) x = -x^t K x \Rightarrow x^t K x = 0$
- Clearly, $x^t A x \in \mathbb{R}$ for all real vectors $x \in \mathbb{R}^n$, and the condition does not imply that A is symmetric as in the complex Hermitian case. (For complex matrix, $A = H + iG$)
- The matrix A satisfies $x^t A x \geq 0$ for all if and only if $(A + A^t)/2$ has only nonnegative eigenvalues. The condition does not automatically imply that A is symmetric as in the complex Hermitian case.
- Every skew-symmetric matrix $K \in M_n(\mathbb{R})$ is orthogonally similar to O_{2k} and

$$\begin{pmatrix} 0 & s_j \\ -s_j & 0 \end{pmatrix}, \quad j = 1, \dots, k,$$

where $s_1 \geq \dots \geq s_k > 0$ are nonzero singular values of A .

General $U^t A U$ form:

Real Normal:

$$A = U \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix} U^t$$

Induction:

$$U^t A U = \begin{bmatrix} \lambda_1 & & \\ 0 & A_1 & \\ & & \ddots \end{bmatrix}$$

$$\downarrow$$

$$\begin{bmatrix} \lambda_1 & & x \\ 0 & & A_k \end{bmatrix}$$

$$A_k - \lambda I =$$

$$u, \bar{u},$$

$$A(x+iy) = u(x+iy)$$

Real symmetric:

$\forall A = A^t$ & A is real, then $A = A^t$.

$\forall A = A^t$ & A is real, then $A = A^t$.

$$U \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix} U^t, \quad U \begin{pmatrix} s_1 & & 0 \\ & \ddots & \\ 0 & & s_n \end{pmatrix} U^t$$

3 Eigenvalues and singular values inequalities

We study inequalities relating the eigenvalues, diagonal elements, singular values of matrices in this chapter.

For a Hermitian matrix A , let $\lambda(A) = (\lambda_1(A), \dots, \lambda_n(A))$ be the vector of eigenvalues of A with entries arranged in descending order. Also, we will denote by $s(A) = (s_1(A), \dots, s_n(A))$ the singular values of a matrix $A \in M_{m,n}$. For two Hermitian matrices, we write $A \geq B$ if $A - B$ is positive semidefinite.

3.1 Diagonal entries and eigenvalues of a Hermitian matrix

The following classical result is useful.

$$\overbrace{x^*Ax}^{\text{first } k \text{ nonzero}} = \overbrace{(x^*Ax)^T}^{\dots} = \overbrace{(x^*Ax)^*}^{\dots} = x^* A^* x = \overbrace{x^*Ax}^{\dots}$$

Lemma 3.1.1 (Rayleigh principle) Let $A \in M_n$ be Hermitian. Then for any unit vector $x \in \mathbb{C}^n$,

$$\lambda_1(A) \geq x^*Ax \geq \lambda_n(A). \quad x^*Ax = x^*U \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix} U^*x \quad \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

The equalities hold at unit eigenvectors corresponding to the largest and smallest eigenvalues of A , respectively.

Proof. Done in homework problem.

$$e_j^t A e_j = \begin{pmatrix} a_{11} & & \\ & \ddots & \\ & & a_{jj} \end{pmatrix} a_{jj} = a_{jj} \quad \square$$

If we take $x = e_j$, we see that every diagonal entry of a Hermitian matrix A lies between $\lambda_1(A)$ and $\lambda_n(A)$.

We can say more in the following. To do that we need the notion of **majorization** and **doubly stochastic matrices**.

A matrix $D = (d_{ij}) \in M_n$ is doubly stochastic if $d_{ij} \geq 0$ and all the row sums and column sums of D equal 1.

Let $x, y \in \mathbb{R}^n$. We say that x is weakly majorized by y , denoted by $x \prec_w y$ if the sum of the k largest entries of x is not larger than that of y for $k = 1, \dots, n$; in addition, if the sum of the entries of x and y , we say that x is majorized by y , denoted by $x \prec y$. We say that x is obtained from y by a pinching if x is obtained from y by changing (y_i, y_j) to $(y_i - \delta, y_j + \delta)$ for two of the entries $y_i > y_j$ of y and some $\delta \in (0, y_i - y_j)$.

Examples

$$\begin{bmatrix} 1 & & & \\ a & 2 & & \\ & & 3 & \\ & & & 100 \end{bmatrix} \quad A = A^* = \begin{bmatrix} 1 & x & x & x \\ x & 2 & x & x \\ x & x & 3 & x \\ x & x & x & 1000 \end{bmatrix} \Rightarrow \lambda_1(A) \geq 1000 \quad \lambda_n(A) \leq 1$$

Can we construct a matrix $A = A^t$ real such that A has eigenvalues $(3, 2, 1)$, diagonal elements $(2, 2, 2)$

Example - D.S. matrices

$$\begin{bmatrix} 0.1 & 0.9 \\ 0.9 & 0.1 \end{bmatrix}_{15} \quad \begin{bmatrix} d & 1-d \\ 1-d & d \end{bmatrix} \quad 0 \leq d \leq 1$$

$n=3$

$$\begin{bmatrix} 0.1 & 0.2 & 0.7 \\ 0.7 & 0.1 & 0.2 \\ 0.2 & 0.7 & 0.1 \end{bmatrix}, \quad \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \quad \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Note:

$A = (a_{ij}) \in M_n$ with $a_{ij} \geq 0$
is d.s. if and only if

$$A \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}$$

and

$$[1 \dots 1] A = [1 \dots 1]$$

Note:

D_1, D_2 are ds.

\Rightarrow ~~$D_1 D_2$~~ $D_1 D_2$ is ds.

Proof

① $D_1 D_2 = (z_{ij}) \Rightarrow z_{ij} > 0$

Let $e = \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}$

& $D_1 D_2 e = D_1 e = e$

$e^t D_1 D_2 = e^t D_2 = e^t$

Examples of

$x \prec_w y$.

$(3, 2, 1) \prec_w (6, 0, 0)$

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$(3, 2, 1) \not\prec_w (4, \frac{1}{2}, 0)$

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$(3, 2, 1) \prec_w (7, 0, 0)$ & $(3, 2, 1) \not\prec (7, 0, 0)$

$$\begin{aligned} (M, 0, \dots, 0) &\rightarrow \left(\frac{M}{2}, \frac{M}{2}, 0, \dots, 0\right) \\ &\rightarrow \left(\frac{M}{n}, \dots, \frac{M}{n}\right) \end{aligned}$$

Theorem 3.1.2 Let $x, y \in \mathbb{R}^n$ with $n \geq 2$. The following conditions are equivalent.

- (a) $x < y$.
- (b) There are vectors x_1, x_2, \dots, x_k with $k < n$, $x_1 = y$, $x_k = x$, such that each x_j is obtained from x_{j-1} by pinching two of its entries.
- (c) $x = Dy$ for some doubly stochastic matrix.

(c) \Rightarrow (a) Suppose $\begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} d_{1j} \\ \vdots \\ d_{ij} \end{pmatrix} \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$

We assume $x_1 \geq \dots \geq x_n$
 $y_1 \geq \dots \geq y_n$.

Otherwise we prove

$$P \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = D Q \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \underbrace{P^t D}_I Q \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

For any $k \in \{1, \dots, n\}$.

let $v_k = \underbrace{[1 \dots 1]_k}_{k} [0 \dots 0]^t$.

Then $\sum_{i=1}^k x_i = v_k^t \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$

$$= v_k^t D \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

$$= [c_1 c_2 \dots c_n] \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

$$= c_1 y_1 + \dots + c_k y_k + \underbrace{c_{k+1} y_{k+1} + \dots + c_n y_n}_{\leq c_n y_k}$$

$$\leq c_1 y_1 + \dots + c_k y_k + c_n y_k$$

$$= \underbrace{c_1 y_1 + \dots + c_k y_k}_{\leq c_k y_k} + \underbrace{c_n y_k}_{(1-c_k) y_k} = y_1 + \dots + y_k$$

$$\leq c_1 y_1 + \dots + c_k y_k$$

$$+ (1-c_1) y_1 + \dots + (1-c_k) y_k = y_1 + \dots + y_k$$

$$P \begin{pmatrix} x \\ 2 \\ 3 \\ 1 \end{pmatrix} \preceq Q \begin{pmatrix} y \\ 0 \\ 6 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix} \preceq D \begin{pmatrix} 6 \\ 0 \\ 0 \end{pmatrix}$$

$$0 \leq c_i \leq 1$$

$$c_1 + \dots + c_n = 1$$

$$= v_k^t \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}$$

$$= v_k^t \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} = k$$

$$k - (c_1 + \dots + c_k) = c_{k+1} + \dots + c_n$$

$$(1-c_1) + \dots + (1-c_k)$$

$$e^t x = e^t D y = e^t y \quad \therefore \sum_{i=1}^n x_i = \sum_{i=1}^n y_i$$

(a) \Rightarrow (b). Consider $y = (y_1, \dots, y_n)^t$
 $x = (x_1, \dots, x_n)^t \dots$

We prove the result by induction on $n \geq 2$.

If $n=2$, $y_1 \geq x_1 \geq x_2 \geq y_2$

$$(x_1, x_2) = (y_1 - d, y_2 + d)$$

$$d = y_1 - x_1$$

Assume $n > 3$. Let k be the max integer so that

$$x_k \geq x_1.$$

Case 1 $^\circ$ If $k=n$,

then $y_n \geq x_1 \geq \dots \geq x_n \geq y_n$. $\therefore x_1 = \dots = x_n$

and
$$\frac{y_1 + \dots + y_n}{n} = \frac{x_1 + \dots + x_n}{n} = x_n$$

$$\Rightarrow y_1 = \dots = y_n = x_n = \dots = x_1$$

Case 2 $^\circ$ If $k < n$,

then change (y_k, y_{k+1}) to $(x_k, \underbrace{y_k + y_{k+1} - x_k}_{y_{k+1}})$

Then we get:

$$(x_1, \dots, x_n)^t \quad \Delta \quad (y_1, \dots, y_{k-1}, x_k, \hat{y}_{k+1}, y_{k+2}, \dots, y_n)^t$$

Apply induction $(x_2, \dots, x_n)^t$, $(y_1, \dots, y_{k-1}, \hat{y}_{k+1}, y_{k+2}, \dots, y_n)^t$

Clearly, the sum of entries of the two vectors in \mathbb{R}^{n-1} are the same.

Case 1 $^\circ$ For $l \leq k$.

$$x_2 + \dots + x_l \leq y_1 + \dots + y_l$$

Case 2 $^\circ$ For $k < l$

$$x_2 + \dots + x_l \leq y_1 + \dots + y_{k-1} + \hat{y}_{k+1} + y_l = y_1 + \dots + y_l$$

Theorem 3.1.3 Let $\mathbf{d}, \mathbf{a} \in \mathbb{R}^n$. The following are equivalent. *order of entries does not matter*

- (a) There is a complex Hermitian (real symmetric) $A \in M_n$ with entries of \mathbf{a} as eigenvalues and entries of \mathbf{d} as diagonal entries.
- (b) The vectors satisfy $\mathbf{d} \prec \mathbf{a}$.

$$A = \begin{pmatrix} d_1 & & * \\ & \ddots & \\ * & & d_n \end{pmatrix}$$
 has e.v. (a_1, \dots, a_n) ,
 then $S \begin{pmatrix} d_1 & & * \\ & \ddots & \\ * & & d_n \end{pmatrix} S^{-1} = \begin{pmatrix} a_1 & & * \\ & \ddots & \\ * & & a_n \end{pmatrix}$
 has e.v. a_1, \dots, a_n .
 For any order d_{i_1}, \dots, d_{i_n}
 PAP^T $= \begin{pmatrix} d_{i_1} & & * \\ & \ddots & \\ * & & d_{i_n} \end{pmatrix}$

Remark If $X \prec Y$.
 then the sum of the k smallest entries of X
 ~~$\leq \sum_{i=1}^k x_i$~~
 \leq k smallest entries of Y

$$S = \sum_{i=1}^n x_i = \sum_{i=1}^n y_i$$

$$\sum_{i=k+1}^n x_i = S - \sum_{i=1}^k x_i$$

$$\sum_{i=k+1}^n y_i = S - \sum_{i=1}^k y_i$$